

New Tools of Phantoms, Monte Carlo Person-Specific Organ Dosimetry in Radiation Protection: Do We Have the Necessary Computational Tools for a Paradigm Change?

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Disclosures:



Developer of commercial software tools:



By Virtual Phantoms <u>https://www.virtual-dose.com</u>







By Wisdom Tech http://www.wisdom-tech.online/

Learning objectives:

1. To gain a historical perspective of computational human phantoms and Monte Carlo dose calculations;

2.To learn about the latest research and development on patient-specific phantoms involving automatic multi-organ segmentation tools;

3.To learn about the latest research and development on rapid Monte Carlo dose calculation tools involving GPU co-processors as well as virtual-sourcemodeling of CT and PET/CT scanner, radiotherapy linac, and those found in common radiation protection environments.

60-Year History of Computational Phantoms

SCHEES IN MEDICAL PRESIDES AND BIOMEDICAL ENGINEERING

HANDBOOK OF

ANATOMICAL MODELS

FOR RADIATION

DOSIMETRY

Xie George Xu and Keith F. Eckerman

- Radiation Protection
- Medical Imaging
- Radiotherapy

1st Generation STYLIZED

2nd Generation VOXEL

3rd Generation <u>BREP</u>

Biological and Medical Physics, Biomedical Engineering

The Phantoms

of Medical and

Health Physics

Devices for Research and Development

D Springer

Larry A. DeWerd

Michael Kissick Editors

	IOP Publishing Institute of Physics and Engineering in Medicine	Physics in Medicine & Biology
-	Phys. Med. Biol. 59 (2014) R233R302	doi:10.1088/0031-9155/59/18/R233
	Topical Reviews	

An exponential growth of computational phantom research in radiation protection, imaging, and radiotherapy: a review of the fifty-year history

X George Xu

ORNL family models 1960-1980s

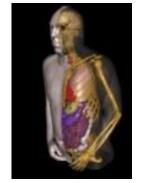
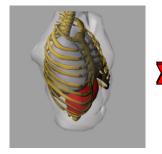


Image-based 1980s-present



PersonalizationMulti-scale

(voxel – DNA)

Deformable 4D models 2000s-present

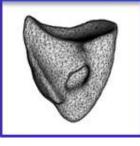
Comments

- For decades, in imaging procedures, all "organ doses" were assessed using ICRP "Reference Man" approach involving "population-averaged phantom libraries"
- "Patient-specific organ doses" would require <u>two newly available tools</u>:
 - 1. automatic organ segmentation
 - 2. "near real-time" Monte Carlo simulations

Outline of the presentation

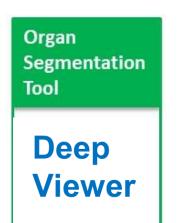


Anatomical Modeling Tool





VirtualDose



Dose Computing Tool

Two Ways to Determine Organ Doses

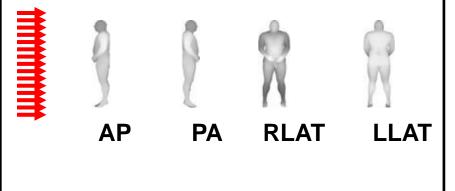
Measurements

DosimetersPhysical phantom



Monte Carlo Simulations

- Computational phantoms
- Monte Carlo codes



60-Year History of Computational Phantoms

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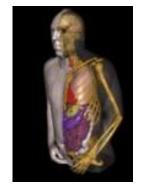
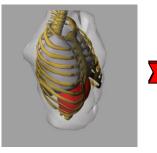


Image-based 1980s-present



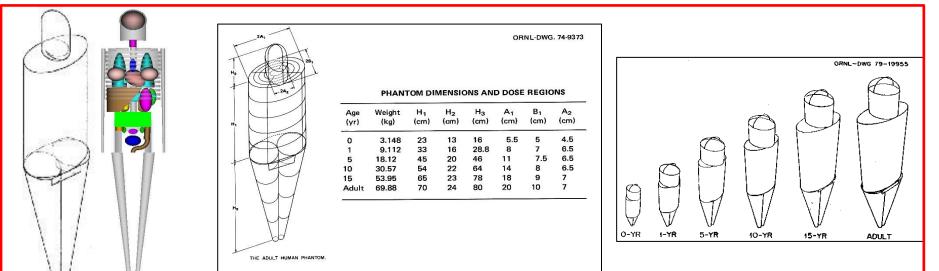
PersonalizationMulti-scale

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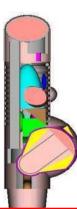
Deformable 4D models 2000s-present

1st-Generation "Stylized" Phantoms (Society of Nuclear Medicine's MIRD Committee)

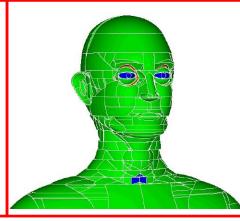
- ICRP paradigm based on "population averaged" Reference Man concept
- Anatomically simple and friendly for computers prior to 1980s



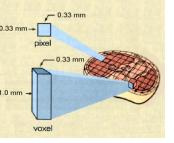
Snyder *et al* (1978), Cristy (1980), and Cristy and Eckerman (1987), plus ADAM/Eva phantoms by Kramer et al (1982) from Germany



At the end of each trimester of pregnancy (**Stabin** *et al* **1995**)



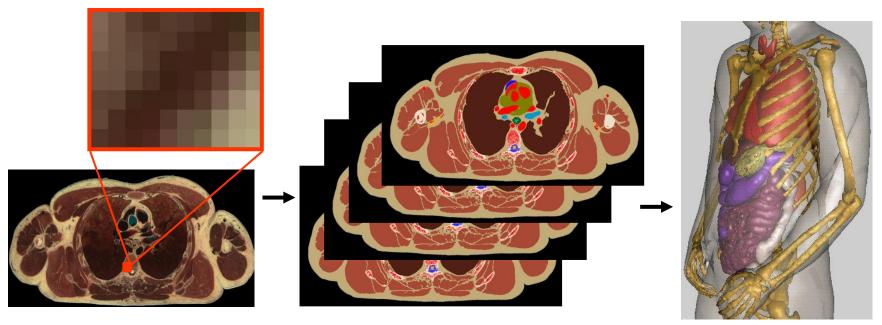
- The Computational Anatomical Man (CAM) and
- CAF (Computerized Anatomical Female)
- by Billings and Yucker (1973)
- Used exclusively by NASA



2nd-Generation "Voxel" Phantoms - Example of the VIP-Man (1997-2000)



Xu XG, Chao TC, Bozkurt A. VIP-Man: An image-based whole-body adult male model constructed from color photographs of the Visible Human Project for multi-particle Monte Carlo calculations. <u>Health Phys</u>., 78(5):476-486, 2000. *One of the most cited (450+)*



Identification of organs in each slice of a 2D pixel map

Registration of all slices

Finished 3D voxel VIP-Man

<<Handbook of Anatomical Models for Radiation **Dosimetry>> by Xu and Eckerman 2009** – curtsey images





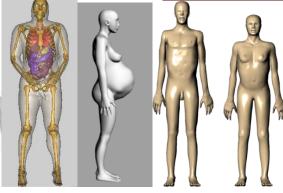


NORMAN

MAX06 FAX06

Zubal

NCAT

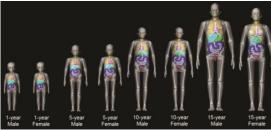


HANDBOOK OF ATOMICAL MODELS FOR RADIATION DOSIMETRY

VIP-Man, Pregnant, Adult M/F







UF Family



Otoko Onago JM KF



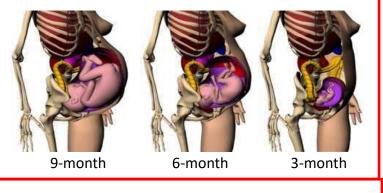
KTMAN 1, 2

CNMAN VCH

3rd-Generation "BREP" Phantoms (NURBS or Meshes)

Pregnant Phantoms

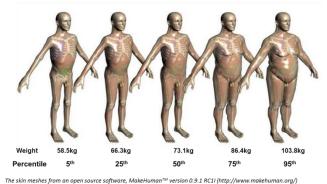
Xu X G, Taranenko V, Zhang J, Shi C. A boundary-representation Inethod for designing whole-body radiation dosimetry models: pregnant females representing three gestational periods, RPI-P3, -P6 and –P9. <u>Phys. Med. Biol</u>. (2007) *The Best 10 papers by PBM in 2007*



• Same height (e.g. 176cm Male), but different weights:

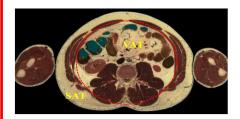
Size and Weight Adjustable Phantoms

Na YH, Zhang^{*} B, Zhang^{*} J, Caracappa PF, Xu XG. Deformable Adult Human Phantoms for Radiation Protection Dosimetry: Anatomical Data for Covering 5th- 95th Percentiles of the Population and Software Algorithms. <u>Phys. Med. Biol</u>. 55: 3789-3811 (2010).



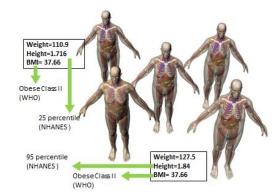
Obese Individuals

Ding A, Mille MM, Liu T, Caracappa PF and Xu XG. <u>Phys. Med. Biol</u>. 57:2441–2459 (2012). *One of the most downloaded in 2012*

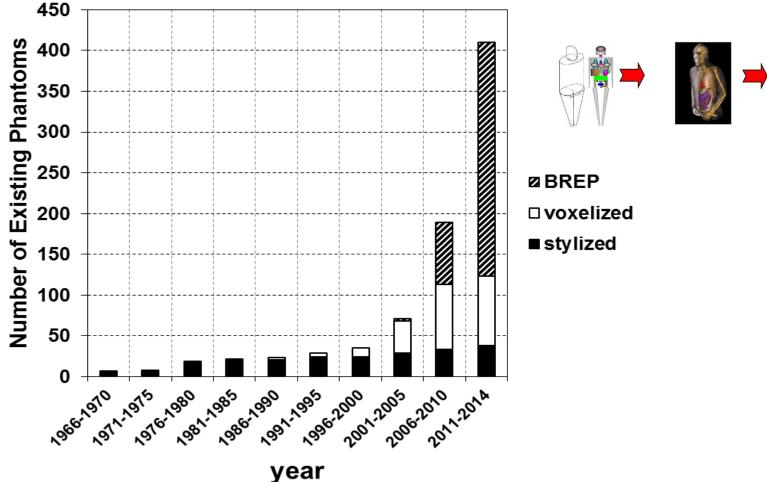


Weight Category	ВМІ
Underweight	< 18.5
Normal Weight	18.5-24.9
Overweight	25-29.9
Obese (I, II)	30-34.9, 34.9-39.9
Morbidly Obese	>40

Obese Phantoms

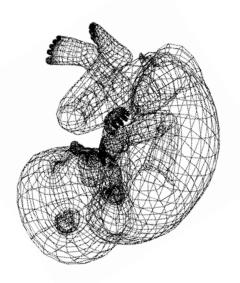


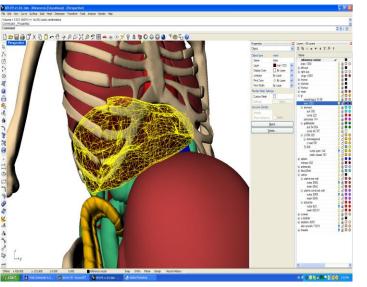
The number of phantoms in existence since 1966 shows a surprising exponential growth



X. George Xu, "An exponential growth of computational phantom research in radiation protection, imaging, and radiotherapy: a review of the fifty-year history," <u>Physics in Medicine & Biology</u>, 59(R233-R302) 2014.

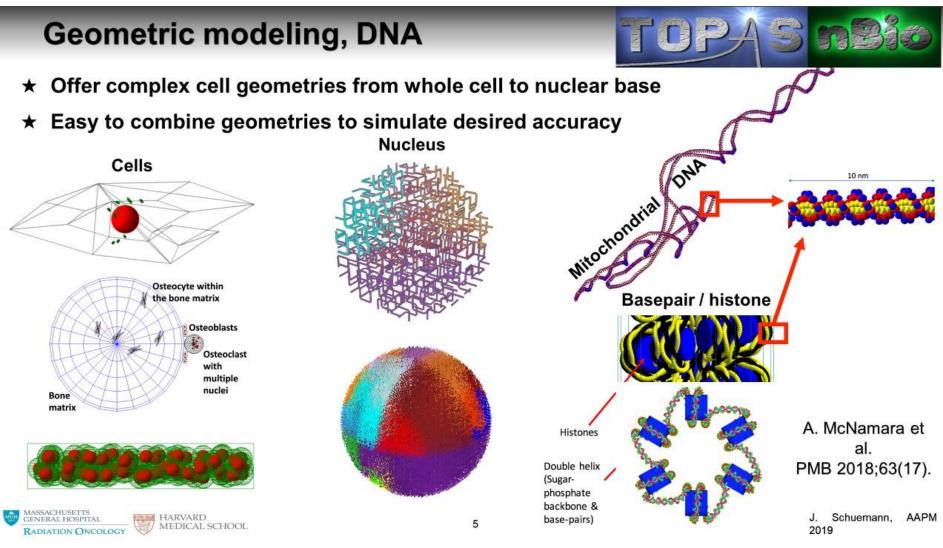
Voxel mesh geometric modeling tools





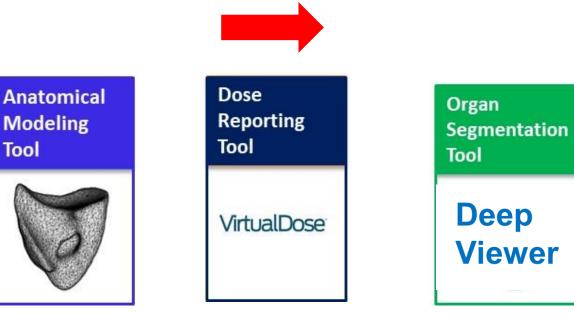


Future Radiobiological Predictive Modeling at Multi-Scale - To Bridge the Gap Between Voxel and DNA



Courtesy slide from Jan Schuemann

Outline of the presentation



Tool



Cancer risks of pediatrics from CT by Pearce, et al. Lancet 2012

- 178,604 young patients
- CT scans from 1985 -2002
- 81 hospitals in Great Britain
- Cancer data from 1985 2008



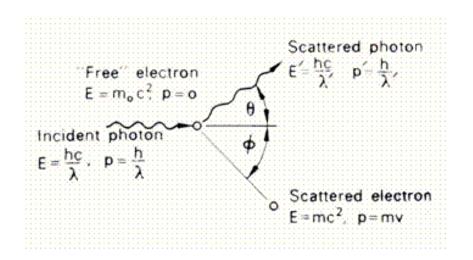
	Male patients		Female patients	
	Brain dose (mGy)	Red bone marrow dose (mGy)	Brain dose (mGy)	Red bone marrow dose (mGy)
Age at brain CT				
0 years	28	8	28	8
5 years	28	9	28	9
10 years	35	6	35	6
15 years	43	4	44	6
20 years	35	2	42	2
Age at chest CT				
0 years	0.4	4	0-4	4
5 years	0.3	3	0-3	3
10 years	0.3	3	0-3	3
15 years	0.2	4	0-3	4
20 years	0.2	4	0-3	4
Age at abdominal CT				
0 years	0.2	3	0.2	3
5 years	0.1	2	0.1	2
10 years	0.1	3	0.1	3
15 years	0.0	3	0-0	3
20 years	0.0	3	0-0	4
Age at extremity CT				
0 years	0.0	1	0-0	1
5 years	0.0	0.2	0-0	0-2
10 years	0.0	0.1	0-0	0.1
15 years	0.0	0.0	0-0	0-0
20 years	0.0	0.0	0-0	0.0

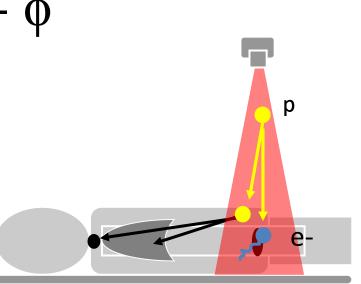
Table 1: Estimated radiation doses to the brain and red bone marrow from one CT scan, by $\frac{1}{5}$ can type, sex, and age at scan, as used in this study for scans after 2001

Radiation Physics for CT Scans

X-ray photon interactions (< 160 keV)</p> ✓ Photoelectric effect







Method: Modeling methodology

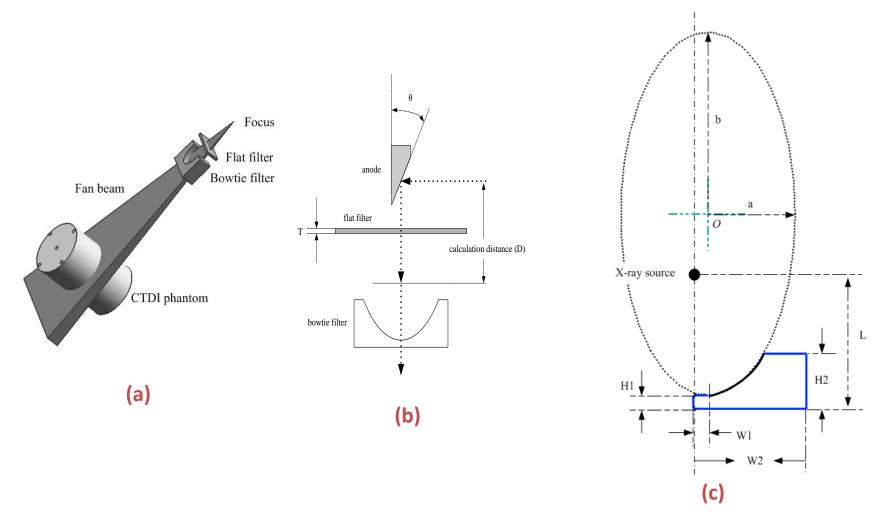


Illustration of components of CT scan and modeling of CT scan, (a) components of the scan, (b) illustration of CT source model, (c). illustration of parameters for BTF (Gu et al. 2009)



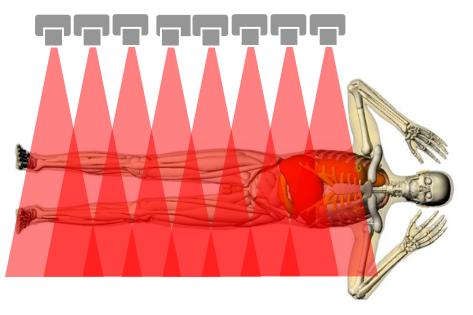
A Comprehensive Slice-byslice Organ Dose Database

- Axial scan simulations in MCNPX
- Contiguous scans from the top to the bottom of <u>27 phantoms</u>
- CT technical parameters

bowtie filters

-4 different tube voltages: 80, 100,
120, and 140 kVp
-4 different beam collimations: 1.25
mm, 5 mm , 10 mm and 20 mm
-Using both the head and body

http://www.virtual-dose.com



Database Hosted in SQL Data Server

Organ name

Kicrosoft SQL Server Management Studio				-	and the owner of the	-	-	the state of the	-		-	-			_ 0 <u>_ x</u>
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🕀 💷 dbo.RPI_P6_Body_5mm_100kvp	Adrenals	3.8E-10	2.64E-10	3.35E-10	3.6E-10	4.01E-10	3.09E-10	4.29E-10	3.64E-10	3.07E-10	2.7E-10	6.02E-10	5.4E-10	6.64E-10	5.85E-10
	Bladder_wall	4E-09	5.04E-09	5.28E-09	4.91E-09	5.4E-09	5.35E-09	4.81E-09	5.25E-09	5.08E-09	5.31E-09	5.36E-09	5.41E-09	5.59E-09	5.92E-09
	Brain	6.44E-11	9.77E-11	7.37E-11	7.42E-11	1.03E-10	1.13E-10	8.91E-11	1.04E-10	1.03E-10	1.31E-10	1.06E-10	1.04E-10	9.46E-11	8.99E-11
	Breasts	6.84E-10	6.97E-10	7.1E-10	7.27E-10	7.3E-10	7.72E-10	7.95E-10	7.71E-10	7.99E-10	8.2E-10	8.48E-10	8.84E-10	9.38E-10	8.84E-10
	Esophagus	2.14E-10	3.58E-10	2.4E-10	1.54E-10	1.74E-10	2.34E-10	2.56E-10	1.71E-10	2.69E-10	2.97E-10	2.49E-10	2.79E-10	3.02E-10	3.17E-10
	Eye_lens	9.26E-10	4.41E-10	7.68E-10	1.11E-09	7.7E-10	1.41E-09	1.73E-09	0	1.29E-09	1.31E-09	5.12E-10	3.14E-10	0	1.45E-10
H dbo.RPI_P6_Head_10mm_140kvp	Eyeballs	3.67E-10	3.04E-10	2.89E-10	3.66E-10	4.15E-10	2.93E-10	3.44E-10	2.53E-10	4E-10	2.73E-10	3.46E-10	4.13E-10	3.19E-10	4.48E-10
	Fetal_brain	2.02E-09	2.35E-09	2.59E-09	2.37E-09	2.39E-09	2.58E-09	2.21E-09	2.43E-09	2.55E-09	2.54E-09	2.58E-09	2.7E-09	2.57E-09	2.95E-09
	Fetal_skeleton	7.67E-09	9.72E-09	1.01E-08	9.92E-09	9.77E-09	9.94E-09	9.26E-09	9.44E-09	9.58E-09	9.37E-09	9.64E-09	9.93E-09	1E-08	1.03E-08
	Fetal_soft_ti	2.68E-09	3.52E-09	3.71E-09	3.6E-09	3.48E-09	3.4E-09	3.36E-09	3.24E-09	3.32E-09	3.32E-09	3.43E-09	3.55E-09	3.52E-09	3.53E-09
H dbo.RPI P6 Head 20mm 140kvp	Fetus_total	2.6E-09	3.38E-09	3.58E-09	3.45E-09	3.35E-09	3.3E-09	3.22E-09	3.15E-09	3.23E-09	3.22E-09	3.33E-09	3.45E-09	3.41E-09	3.46E-09
H dbo.RPI_P6_Head_20mm_80kvp	Gallbladder	2.84E-10	3.29E-10	3.16E-10	2.63E-10	3.42E-10	4.42E-10	3.89E-10	3.33E-10	4.58E-10	5.43E-10	4.27E-10	3.94E-10	5.77E-10	3.58E-10
	Heart_wall	2E-10	2.76E-10	2.07E-10	3.02E-10	2.93E-10	2.45E-10	2.73E-10	2.95E-10	3.1E-10	3.73E-10	2.63E-10	3.96E-10	3.5E-10	3.62E-10
	Kidneys	5.04E 10	5.325 10	4.9E 10	5.44E 10	5.03E 10	4.93E 10	E.08E 10	5.65E 10	5.145 10	E.00E 10	6.4E 10	C.0E 10	6.36E 10	7.025 10
	LI_conts	1.45E-09	1.6E-09	1.69E-09	1.74E-09	1.82E-09	1.81E-09	1.76E-09	1.73E-09	1.81E-09	1.84E-09	1.99E-09	1.98E-09	2.06E-09	1.99E-09
H dbo.RPI P6 Head 5mm 80kvp	LI_wall	1.46E-09	1.63E-09	1./E-09	1./1E-09	1.//E-09	1./6E-09	1./3E-09	1.78E-09	1./5E-09	1.82E-09	1.89E-09	1.96E-09	1.98E-09	2.04E-09
	Liver	3.03E-10	3.42E-10	3.16E-10	3.44E-10	3.09E-10	3.45E-10	3.48E-10	4.13E-10	3.39E-10	4.06E-10	4.72E-10	4.15E-10	4.69E-10	4.6E-10
	Lungs	5.51E-10	5.77E-10	6E-10	5.85E-10	6.03E-10	6.31E-10	6.63E-10	6.57E-10	6.67E-10	7.5E-10	J2E-10	7.31E-10	7.73E-10	8.03E-10
	Ovaries	1.4E-09	1.64E-09	1.45E-09	1.57E-09	1.57E-09	1.24E-09	1.28E-09	1.1E-09	1.76E-09	1.58E-09	9E-09	2.02E-09	2.19E-09	2.01E-09
	Pancreas	4E-10	4.44E-10	4.7E-10	5.79E-10	4E-10	3.77E-10	6.72E-10	5.42E-10	4.61E-10	4.36E-10	-10	5.24E-10	6.64E-10	5.26E-10
	Placenta	6.89E-10	8.77E-10	8.7E-10	9.27E-10	8.55E-10	8.41E-10	8.81E-10	8.68E-10	8.52E-10	9.24E-10		9.24E-10	9.23E-10	1E-09
	Remainder	7.46E-09	4.5E-08	5.29E-08	4.37E-08	3.34E-08	2.68E-08	2.41E-08	2.17E-08	2.05E-08	2.25E-08	2.1 0	2E-08	2.16E-08	2.36E-08
	SI_wall_and	6.99E-10	7.29E-10	7.17E-10	7.38E-10	7.95E-10	7.53E-10	7.93E-10	7.95E-10	8.18E-10	8.53E-10	9.39E-10	9E-10	1.01E-09	9.79E-10
	Skeleton	2.58E-08	1.48E-07	3.23E-07	3.97E-07	4.24E-07	3.84E-07	3.03E-07	2.44E-07	2.05E-07	1.52E-07	1.6E-07	1.96E-07	1.72E-07	1.46E-07
	Skin	8.77E-08	2.16E-07	1.24E-07	1.13E-07	9.85E-08	8.57E-08	7.24E-08	6.25E-08	5.41 - 08	3017-08	O S€®r(2 504 5		4.86E-08
	Spleen	5.13E-10	4.43E-10	4.52E-10	4.05E-10	3.49E-10	4.61E-10	4E-10	5.12E-10	4.20L-10	4.20L-10	7.021-10	4.74E-10	5.22E-10	5.01E-10
	Stomach_wall	2.67E-10	3.28E-10	2.4E-10	2.88E-10	2.85E-10	2.91E-10	3.06E-10	3.5E-10	3.47E-10	3.32E-10	4.39E-10	3.99E-10	4.33E-10	4.91E-10
	Thymus	1.46E-10	2.3E-10	2.32E-10	1.87E-10	2.15E-10	2.68E-10	2.2E-10	1.25E-10	1.24E-10	1.09E-10	7.54E-11	1.81E-10	2.23E-10	2.51E-10
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	14 4 1 of 33														

Easy to manage and update

http://www.virtual-dose.com

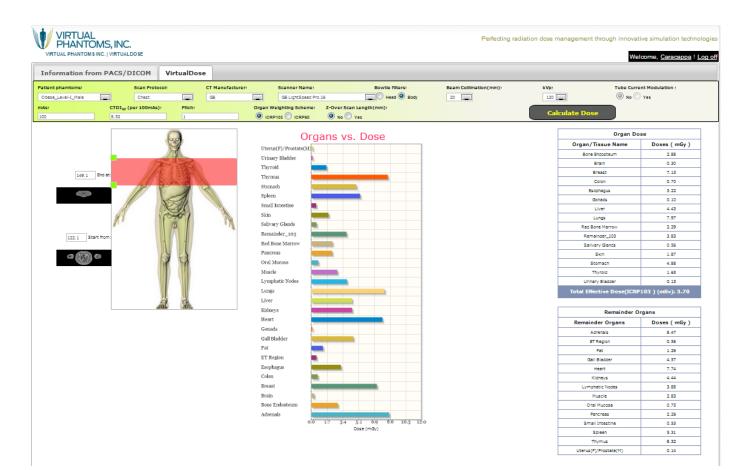
VirtualDose

A product of Virtual Phantoms. Inc

CT Dose Reporting Software

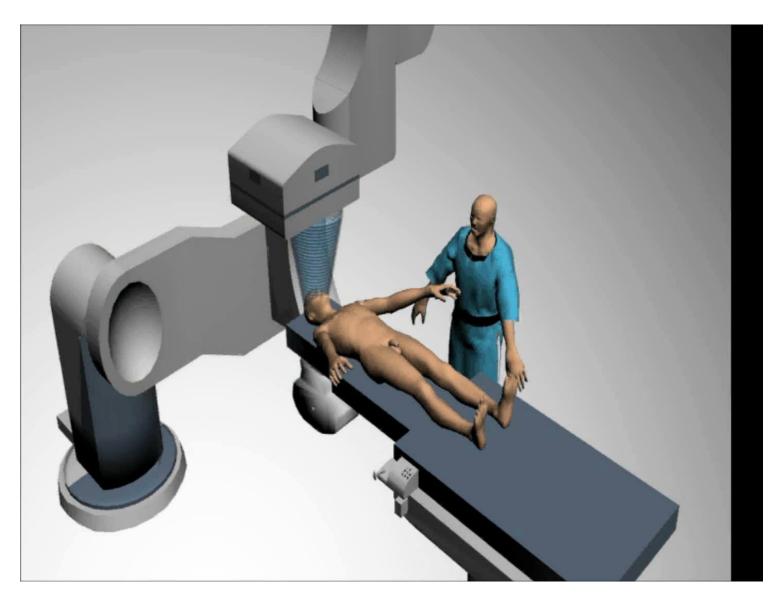
VirtualDose A product of Virtual Phantoms, Inc.

http://www.virtual-dose.com



In 2021-2022, over <u>31.5-million</u> dose calculation requests from more than 275 sites worldwide were processed

VIRTUAL PHANTOMS, INC.

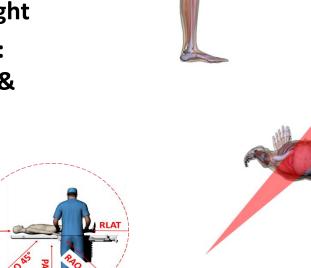


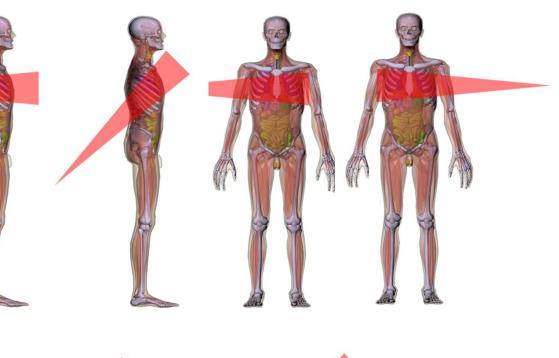


VirtualDose-IR Dose from fluoroscopically guided intervention (FGI) - Patients

- Simulated beam directions:
 - Posterior Anterior, Crani
 45 °
 - Lateral: left & right
 - Oblique:
 45° left & right

LLAT



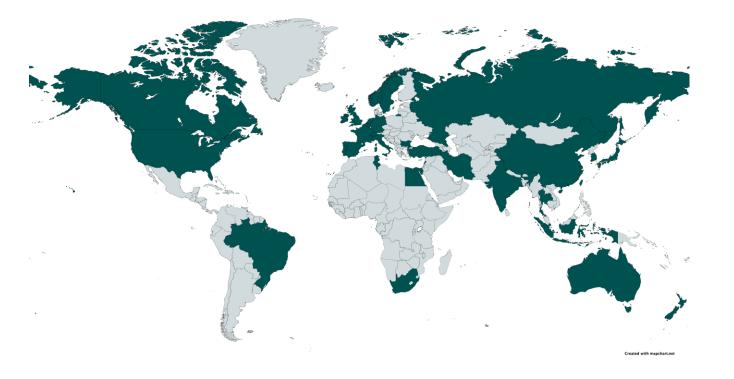






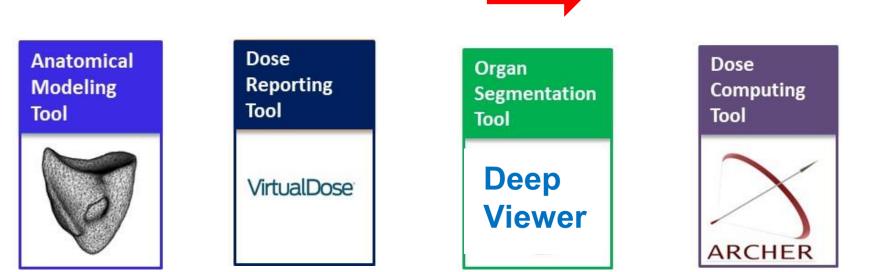
Worldwide Users

A product of Virtual Phantoms, Inc.



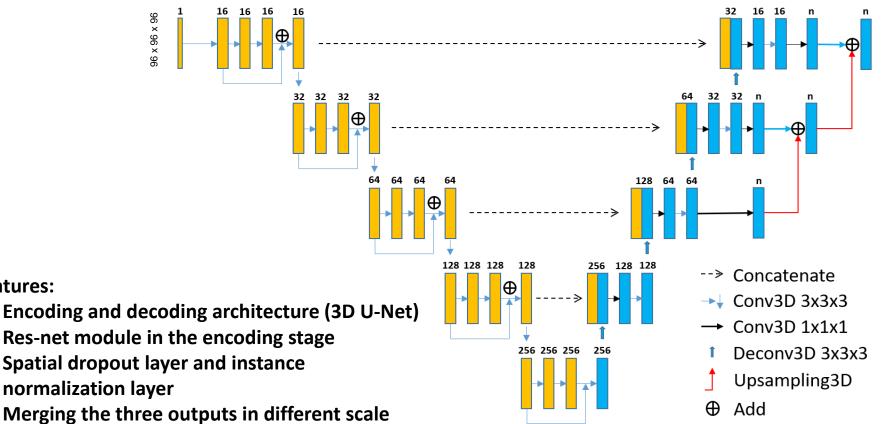
31.5Million+ Dose Calculations Per Year

Outline of the presentation



Moving towards Patient-Specific Phantoms - **DL-based Auto Multi-organ Segmentation**

Network architecture [1]:



Merging the three outputs in different scale and getting the final segmentation results

Features:

[1] O. Ronneberger, P. Fischer, and T. Brox, "U-net: Convolutional networks for biomedical image segmentation," in International *Conference on Medical image computing and computer-assisted intervention*, 2015: Springer, pp. 234-241.

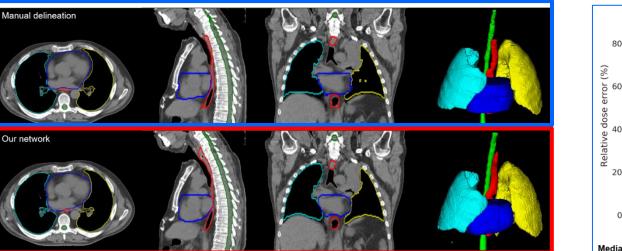
Dataset 1: Lung CT Segmentation Challenge 2017 (LCTSC) [2]

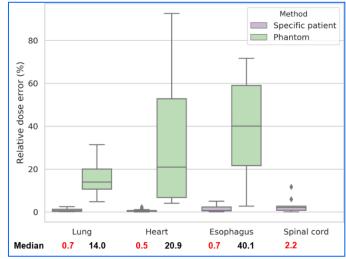
[2] Yang Jinzhong, et al. Data from Lung CT Segmentation Challenge (2017). The Cancer Imaging Archive.

- Training and validation set: 48 patients
- Testing set: 12 patients
- Loss function: weighted dice similarity
- 5 segmented organs: left lung (yellow), right lung (cyan), heart (blue), spinal cord (green), and esophagus (red)

Methods	Left lung	Right lung	Heart	Esophagus	Spinal cord							
Interrater variability	0.96±0.02	0.96±0.02	0.93±0.02	0.82±0.04	0.86±0.04							
1 (0.97±0.02	0.97±0.02	0.93±0.02	0.72±0.10	0.88±0.04							
2 (0.98±0.01	0.97±0.02	0.92±0.02	0.64±0.20	0.89±0.04							
3 (0.98±0.02	0.97±0.02	0.91±0.02	0.71±0.12	0.87±0.11							
4 (0.97±0.01	0.97±0.02	0.90±0.03	0.64±0.11	0.88±0.05							
5 (0.96±0.03	0.95±0.05	0.92±0.02	0.61±0.11	0.85±0.04							
6 (0.96±0.01	0.96±0.02	0.90±0.02	0.58±0.11	0.87±0.02							
7 (0.95±0.03	0.96±0.02	0.85±0.04	0.55±0.20	0.83±0.08							
Ours	0.96±0.01	0.96±0.02	0.93±0.02	0.73±0.10	0.88±0.04							

Dice similarity coefficient (mean ± standard deviation)





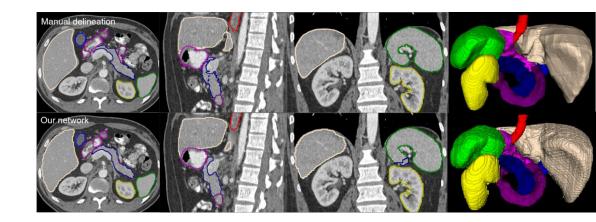
Peng Z, Fang X, Shan H, Liu T, Pei X, Yan P, Wang G, Liu B, Kalra M, Xu XG. Multi-Organ Segmentation of CT Images Using Deep-Learning for Instant and Patient-Specific Dose Reporting. 28 AAPM, San Antonio, TX, July 14 – 18, 2019.

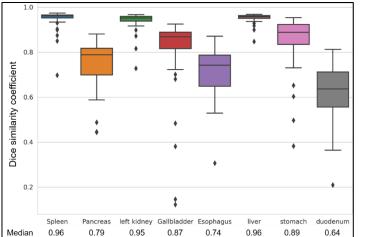
Dataset 2: The Cancer Image Archive (TCIA) Pancreas-CT [3]

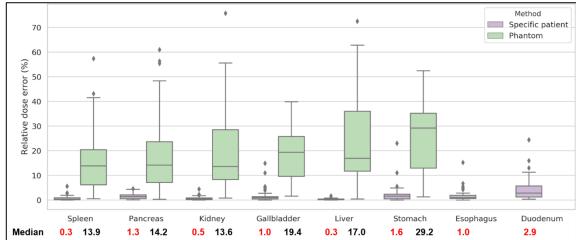
[3] Gibson E, et al. Automatic multi-organ segmentation on abdominal CT with dense v-networks. IEEE Transactions on Medical Imaging, 2018.

Total patients: 43 **5 cross-validation:** 8, 8, 9, 9, 9 **Loss function:** weighted dice similarity coefficient

8 segmented organs: spleen (green), pancreas (navy), left kidney (yellow), gallbladder (blue), esophagus (red), liver (bisque), stomach (magenta), and duodenum (purple)

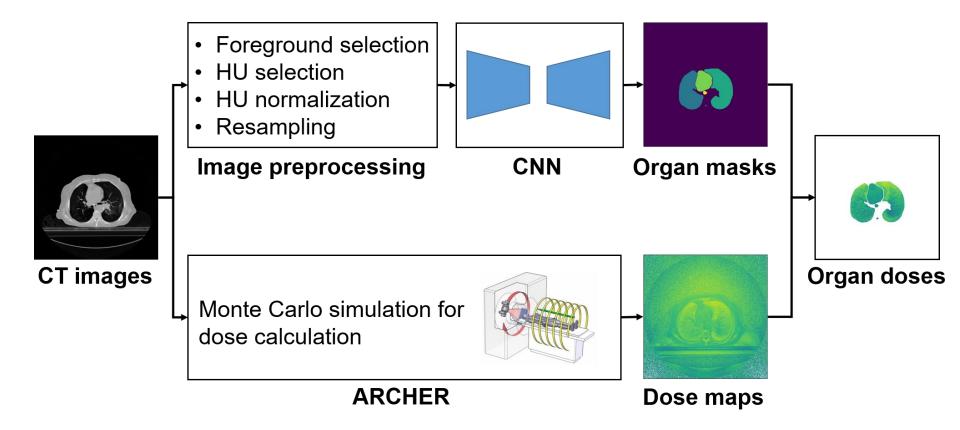




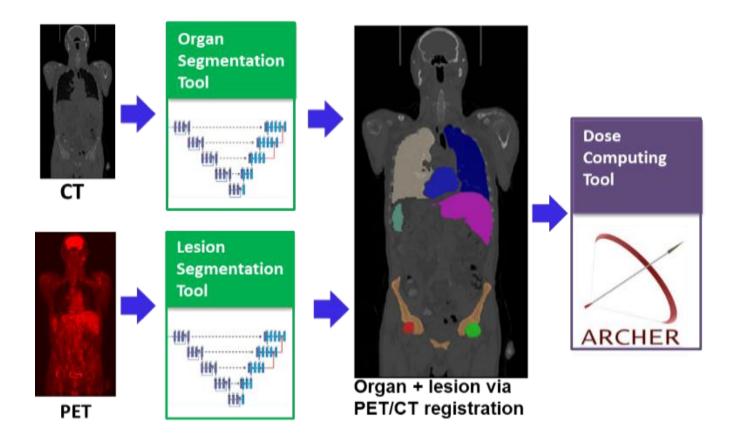


Peng Z, Fang X, Shan H, Liu T, Pei X, Yan P, Wang G, Liu B, Kalra M, Xu XG. Multi-Organ Segmentation of CT Images Using Deep-Learning for Instant and Patient-Specific Dose Reporting. <u>AAPM, San Antonio, TX, July 14 – 18, 2019</u>.

Method: The overall flowchart for creating patient-specific phantoms (CT)



Can also be done for PET/CT patientspecific phantoms





Segmentation Testing for the DeepViewer Tool

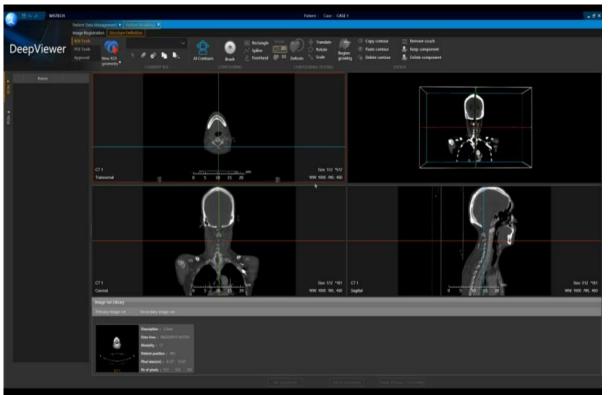
- Can perform more than 40 organs and tissues
- Whole body in 3-5 minutes

Organs	Dice	Organs	Dice	Organs	Dice	Organs	Dice
Skin	0.99	Brainstem	0.94	Stomach	0.88	Larynx	0.85
Brain	0.98	Femoral Head	0.93	Temporal Lobe	0.88	Rectum	0.84
Lung	0.98	Pelvis	0.93	Parotid	0.86	Oral Cavity	0.83
Liver	0.97	Heart	0.93	Esophagus	0.86	Breast	0.81
Spleen	0.95	Mandible	0.93	Pancreas	0.86	Optic Nerve	0.78
Bladder	0.95	Trachea	0.92	Lens	0.85	Pituitary	0.76
Cerebellum	0.95	Spinal Cord	0.92	Thyroid	0.85	Optic Chiasm	0.75
Kidney	0.95	Eye Ball	0.9	Bowel	0.85	Cochlea	0.75

Software Tool for Automatic Target and OAR Contouring in Radiotherapy



DeepViewer

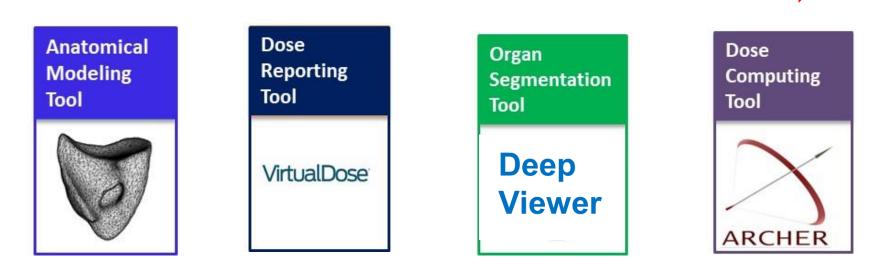




3D visualization

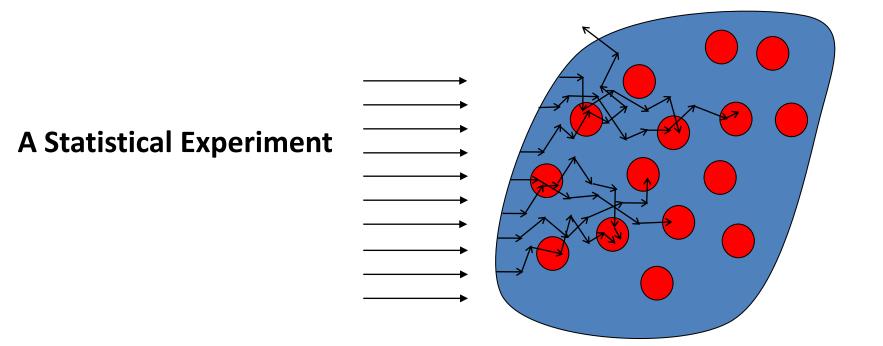
- Support head/neck, chest, abdominal tumor treatment planning
- 40+ OARs
- It takes 3-5 min to complete
- Acceptance rate 95% or better

Outline of the presentation

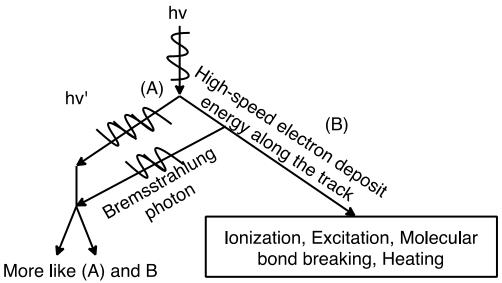


Monte Carlo Methods Ideal for Radiation Transport Simulations, but <u>Used to be Extremely Slow</u>

Boltzmann Transport Equation can be solved by MC methods $\frac{1}{v}\frac{\partial}{\partial t}\psi(\vec{r},\hat{\Omega},E,t) + \hat{\Omega}\cdot\vec{\nabla}\psi(\vec{r},\hat{\Omega},E,t) + \Sigma_{t}(\vec{r},E)\psi(\vec{r},\hat{\Omega},E,t)$ $= \int dE'\int d\Omega'\Sigma_{s}(\vec{r},E'\to E,\hat{\Omega}'\cdot\hat{\Omega})\psi(\vec{r},\hat{\Omega}',E',t) + S(\vec{r},\hat{\Omega},E,t)$



Introduction: Dose Calculation and Monte Carlo Methods



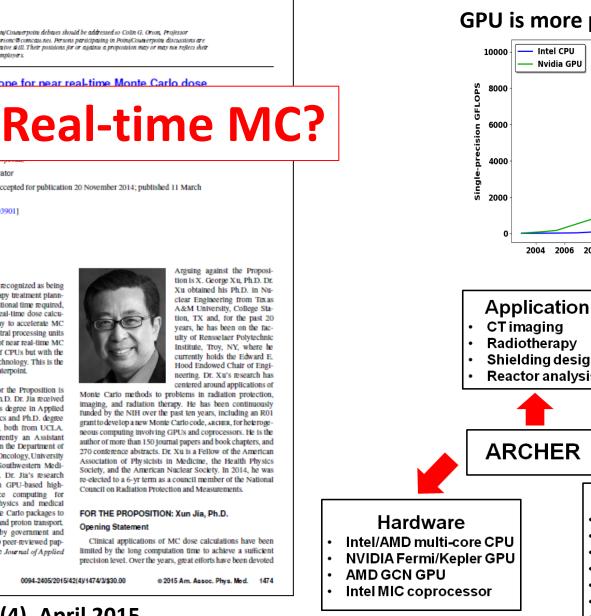
- Dose: energy imparted to matter via ionization and excitation per unit mass
- Deterministic methods subject to non-trivial errors by approximating electron transport in water
- Monte Carlo particle transport is the <u>gold standard</u>
- But lengthy computation time is the main bottleneck of MC applications for clinical settings

The ARCHER Project



- ARCHER (<u>Accelerated Radiation-transport</u> <u>Computations in Heterogeneous</u> <u>EnviRonments</u>)
 - Initiated in 2009
 - Goals:
 - 1. To understand heterogeneous computing architecture and programming models
 - 2. To test code performance on GPUs and MICs
 - 3. To develop functional Monte Carlo codes that take full advantage of CPUs, GPUs and MICs

Solution: ARCHER – A GPU-based Monte Carlo Code



POINT/COUNTERPOINT

Suggestions for topics subable for these Point/Courserpoint debates should be addressed to Colin G. Orton, Professor Emericas, Wayne Scare University, Deutoir, arconc@comcasu.net, Persons participating in Point/Counterpoint discussions are sciecced for heir knowledge and communicative skill. Their positions for or against a proposition may or may not reflect their personal opinions or the positions of their employers.

GPU technology is the hope for near real-time Monte Carlo dose

calculations

Xun Jia, Ph.D. Department of Radiation Oncole (Tel: 214-648-3224; E-mail: xu

X. George Xu. Ph.D. Nuclear Engineering Program, (Tel: 518-276-4014; E-mail: xus

Colin G. Orton, Ph.D., Moderator

(Received 15 November 2014; accepted for publication 20 November 2014; published 11 March 2015)

[http://dx.doi.org/10.1118/1.4903901]

OVERVIEW

Monte Carlo (MC) dose calculations are recognized as being the most accurate modality for radiotherapy treatment planning but, because of the excessive computational time required, they cannot presently be used for near real-time dose calculations. Currently, the most common way to accelerate MC dose calculations is to use clusters of central processing units (CPUs), but some believe that the future of near real-time MC dose calculations lies not with clusters of CPUs but with the use of graphics processing unit (GPU) technology. This is the claim debated in this month's Point/Counterpoint.



Arguing for the Proposition is Xun Jia, Ph.D. Dr. Jia received his Masters degree in Applied Mathematics and Ph.D. degree in Physics, both from UCLA. He is currently an Assistant Professor in the Department of Radiation Oncology, University of Texas Southwestern Medical Center. Dr. Jia's research focuses on GPU-based highperformance computing for medical physics and medical

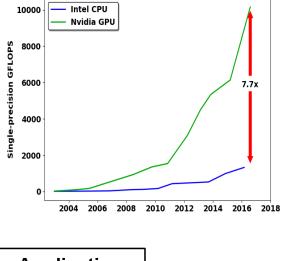
imaging. He has developed several Monie Carlo packages to improve efficiency for photon, electron, and proton transport. Dr. Jia's research has been supported by government and industrial grants and he has published 60 peer-reviewed papers. He is currently a section editor of the Journal of Applied Clinical Medical Physics.

1474 Med. Phys. 42 (4), April 2015

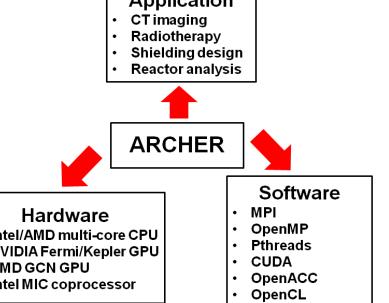
0094-2405/2015/42(4)/1474/3/\$30.00

Med. Phys. 42 (4), April 2015

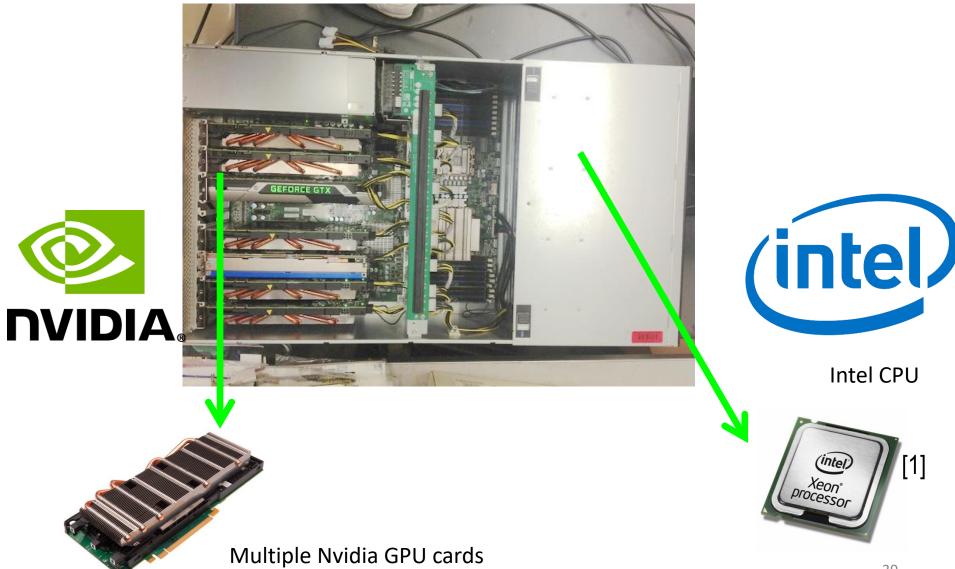
GPU is more powerful than CPU



Cilk



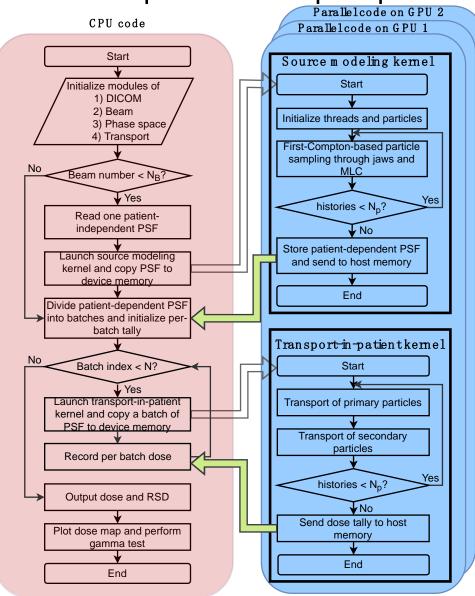
Desk-Top CPU/GPU Hardware @ \$10k



ARCHER Software Development

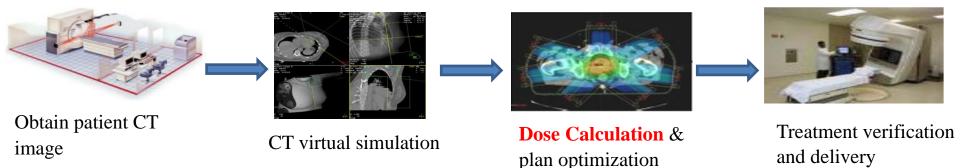
• Workflow of synchronized source model-particle transport process

- Two code variants
- ARCHER_{CPU}: supports multi-threaded parallelization
- ARCHER_{GPU}: supports multi-GPUs parallelization

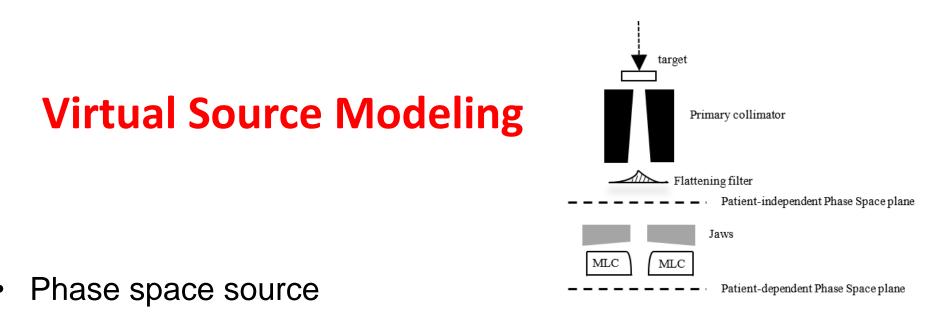


Introduction: External-Beam Radiotherapy

- 1.7 million new cancer cases in the U.S in 2017^[1]
- 52% cancer patients receive external photon beam radiotherapy
- Radiation Therapy aims to maximizes the dose to the tumor, while minimizing the dose to adjacent health tissues



- The accuracy of dose calculations is crucial to treatment planning and dose delivery
 - Accuracy of the source term modeling
- Accuracy of the dose calculation methods
 [1] Siegiel, R., K. D. Miller, and A. Jemal. CA Cancer J Clin 67 (2017): 7-30.

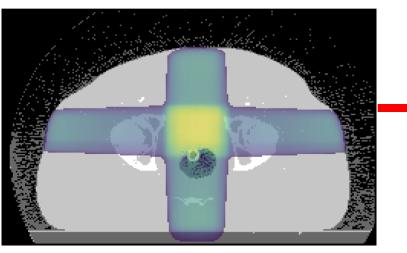


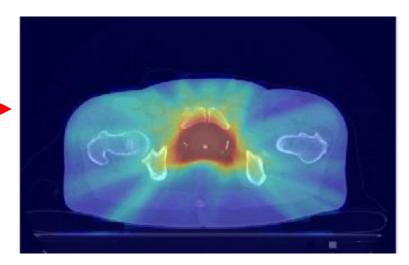
- Phase space particles: a collection of representative pseudo-particles emerging from a radiotherapy source along with their properties (x, y, z, u, v, w, E, wt)
- can be categorized as
 - Patient-independent phase space: could be utilized \succ repeatedly
 - Patient-dependent phase space: specific to each plan and beam 42

Results: ARCHER for External-Beam Therapy (Tomotherapy, IMRT and VMAT) ^[1-5]

Step and Shoot IMRT

VMAT

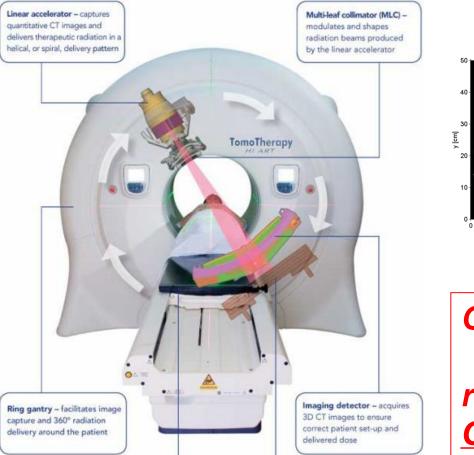


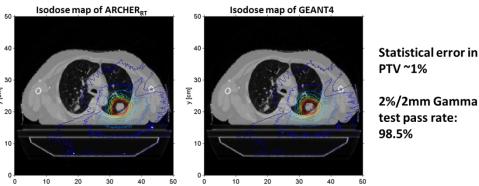


- (1) Liu T. Development of ARCHER a parallel Monte Carlo radiation transport code -- for X-ray CT dose calculations using GPU and coprocessor technologies, Ph.D. <u>Dissertation</u>, Rensselaer Polytechnic Institute (2014).
- (2) Su L. Development and Application of a GPU-Based Fast Electron-Photon Coupled Monte Carlo Code for Radiation Therapy, Ph.D. <u>Dissertation</u>, Rensselaer Polytechnic Institute (2014).
- (3) Lin H. GPU-based Monte Carlo Source Modeling and Simulation for Radiation Therapy involving Varian Truebeam Linac., Ph.D. <u>Dissertation</u>, Rensselaer Polytechnic Institute (2018).
- (4) Su L, Yang YM, Bednarz B, Sterpin E, Du X, Liu T, Ji W, Xu XG. ARCHER_{RT} A Photon-Electron Coupled Monte Carlo Dose Computing Engine for GPU: Software Development of and Application to Helical Tomotherapy. Med Phys. 41:071709 (2014).
- (5) Adam DP*, Liu T, Caracappa PF, Bednarz BP, Xu XG. New capabilities of the Monte Carlo dose engine ARCHER-RT: clinical validation of the Varian TrueBeam machine for VMAT external beam radiotherapy. Med Phys 2537₇ 2549 (2020).

Results: Tomotherapy

Su L, Yang YM, Bednarz, Edmond Sterpin, Du X, Liu T, Ji W, Xu XG. ARCHERRT — A Photon-Electron Coupled Monte Carlo Dose Computing Engine for GPU: Software Development of and Application to Helical Tomotherapy. **Med Phys** (2014).

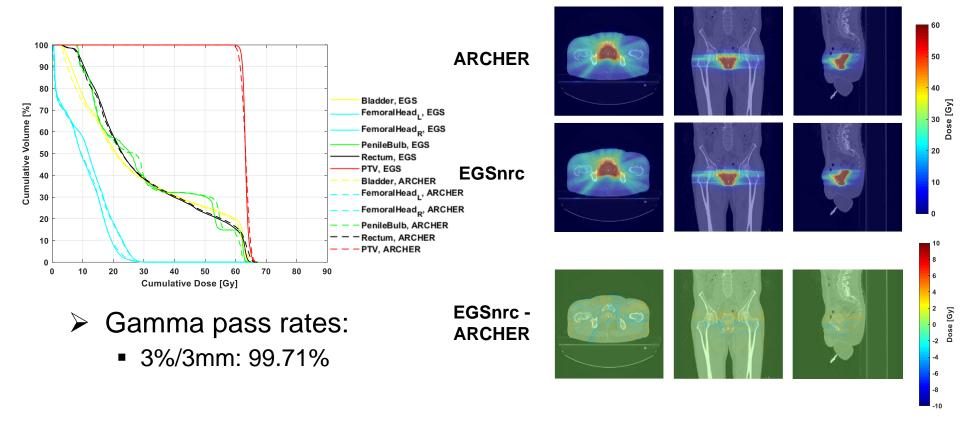




GPU computing time ~ 1 sec,

reducing from GEANT4 ~ 500 <u>CPU hours</u>

Results: Clinical VMAT (prostate case)



ARCHER (NVIDIA 1080Ti GPU) 48 seconds

- Patient transport
- □ Source modeling (linac)

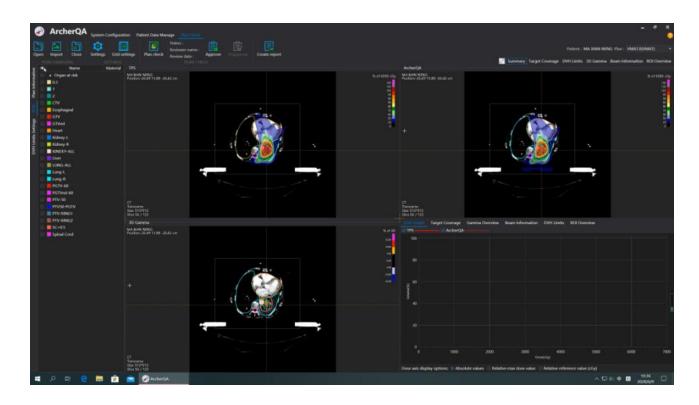
EGSnrc (120-node cluster Intel Xeon E5335 CPUs) 24 hours

Software Tool for 3D Independent Dose-check (verification of TPS results)





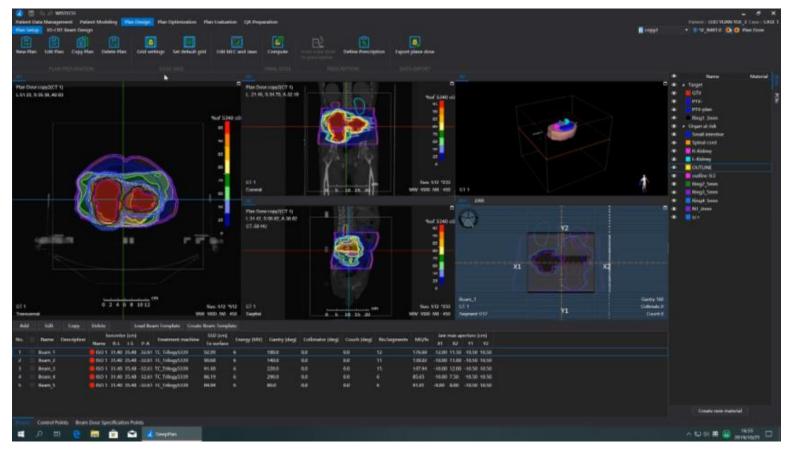
- **1. GPU-accelerated Monte Carlo dose engine**
- 2. Data analyze tools, such as 2D and 3D Gamma Test and DVHs
- **3. TPS and Machine Check**
- 4. Friendly user interfaces, easy to commission and setup



Archer QA GUI

Treatment Planning System (TPS), DeepPlan

- Photons, electrons, protons
- Different dose algorithms including GPU-based Monte Carlo
- Integrated with auto contouring and registration



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- 1. Personalized patient organ-dose always needed for radiation treatment planning
- 2. The workflow greatly facilitated by:
 - AI –based automatic segmentation tools are well tested
 - ✓ Real-time MC computing provides RT QA solutions
- **3.** A paradigm change from "population averaged" Reference Man to "<u>personalized organ doses</u>" is feasible
 - Contact info : <u>xgxu@ustc.edu.cn</u> University of Science and Technology of China





